

1 Synthesis of urban CO₂ emission estimates from multiple methods from the Indianapolis

2 Flux Project (INFLUX)

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4 Jocelyn C. Turnbull^{1,2,*}, Anna Karion³, Kenneth J. Davis⁴, Thomas Lauvaux⁴, Natasha L. Miles⁴,

5 Scott J. Richardson⁴, Colm Sweeney⁵, Kathryn McKain^{2,5}, Scott J. Lehman⁶, Kevin R. Gurney⁷,

6 Risa Patarasuk⁷, Jianming Liang⁷, Paul B. Shepson⁸, Alexie Heimburger⁸, Rebecca Harvey⁸,

7 James Whetstone³

8

9 ¹GNS Science, Rafter Radiocarbon Laboratory, Lower Hutt, New Zealand

10 ²CIRES, University of Colorado, Boulder, CO, USA

11 ³NIST, Gaithersburg, MD, USA

12 ⁴Pennsylvania State University, State College, PA, USA

13 ⁵NOAA/ESRL, Boulder, CO, USA

14 ⁶INSTAAR, University of Colorado, Boulder, CO, USA

15 ⁷Arizona State University, Tempe, AZ, USA

16 ⁸Purdue University, West Lafayette, IN, USA

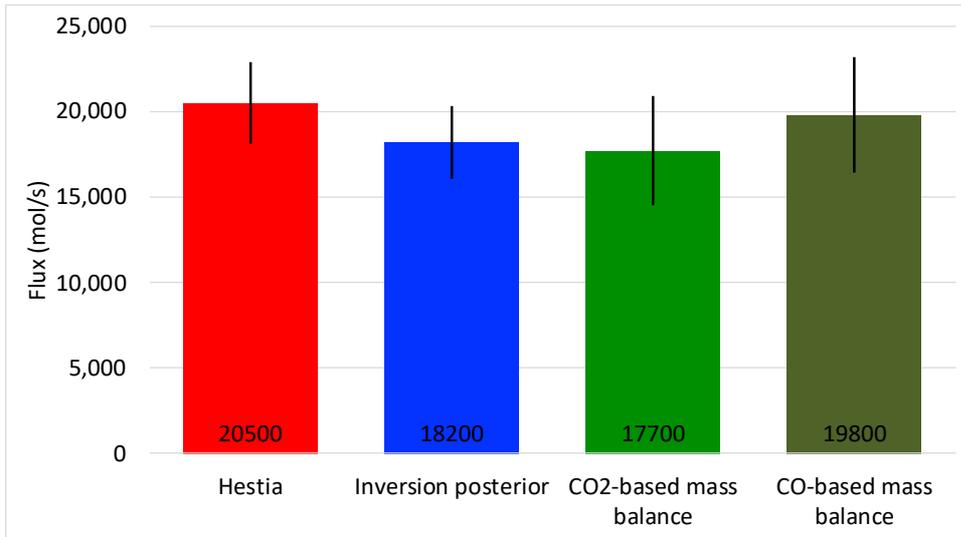
17 *Rafter Radiocarbon Laboratory, GNS Science, 30 Gracefield Rd, Lower Hutt, New Zealand;

18 j.turnbull@gns.cri.nz

19 **Abstract**

20 Urban areas contribute approximately three-quarters of fossil fuel derived CO₂ emissions,
21 and many cities have enacted emissions mitigation plans. Evaluation of the effectiveness of
22 mitigation efforts will require measurement of both the emission rate and its change over
23 space and time. The relative performance of different emission estimation methods is a
24 critical requirement to support mitigation efforts. Here we compare results of CO₂ emissions
25 estimation methods including an inventory-based method and two different top-down
26 atmospheric measurement approaches implemented for the Indianapolis, Indiana, USA
27 urban area in winter. By accounting for differences in spatial and temporal coverage, as
28 well as trace gas species measured, we find agreement among the wintertime whole-city
29 fossil fuel CO₂ emission rate estimates to within 7 %. This finding represents a major
30 improvement over previous comparisons of urban-scale emissions, making urban CO₂ flux
31 estimates from this study consistent with local and global emission mitigation strategy
32 needs. The complementary application of multiple scientifically-driven emissions
33 quantification methods enables and establishes this high level of confidence and
34 demonstrates the strength of the joint implementation of rigorous inventory and
35 atmospheric emissions monitoring approaches.

36



37

38

39 Introduction

40 Urban areas comprise only 3 % of Earth's surface area, but account for ~70 % of global fossil
41 fuel derived carbon dioxide (CO₂ff) emissions ¹. Cities are leading the way in efforts to
42 reduce emissions, with many cities having specific goals for emissions reductions (e.g.
43 c40.org, globalcovenantofmayors.org). Under the Lima-Paris Action Agenda of the Paris
44 Agreement, cities have a formalized role in mitigation strategies ², and indeed many
45 national mitigation objectives will be implemented by local governments and cities.
46 Moreover, policy actions for low-carbon activities and carbon mitigation often provide
47 additional benefits that are important to cities, such as reduced traffic congestion, improved
48 air quality, noise reduction, reduced dependence on imported fuels and potentially
49 improved quality of life and associated economic growth ³.

50

51 In order to evaluate the efficacy of low-carbon and greenhouse gas (GHG) strategies, cities
52 will require continually updated knowledge of GHG emission rates. Ideally, emission
53 information will have the precision and temporal resolution sufficient to evaluate emission
54 trends through time, as well as information about the specific emission source sectors and
55 spatial patterns of emissions within the urban area⁴. Two key questions are: what is the
56 magnitude of whole-city CO₂ emissions; and what is the associated level of uncertainty?
57 City-wide emission estimates have traditionally been obtained using inventory-based
58 methodologies, often adopted from the international Intergovernmental Panel on Climate
59 Change (IPCC) approach and downscaled to the urban scale ⁵⁻⁶. However, a review of
60 traditional urban inventories reveals a mixture of methods and data sources making it
61 difficult to compare cities or assess accuracy or consistency over time⁷⁻⁹, although efforts to
62 standardize methodologies are underway¹⁰⁻¹¹. Uncertainties in these city-scale emission

63 estimates may be 50 to 100 %¹²⁻¹³, insufficient to evaluate emission reduction policies.
64 Under the Paris Agreement Nationally Determined Contributions, nations have committed
65 to emission reductions of about 12 % by 2030 relative to extrapolation of current policy
66 initiatives¹⁴, and many cities propose to reduce their emissions by 30 - 50 % over the next 20
67 to 40 years, equating to 7.5 – 25% decreases per decade (e.g. c40.org; data.cdp.net). These
68 goals suggest that the ability to evaluate urban emissions with an uncertainty of 10 % or less
69 will be needed to provide meaningful assessments of progress.

70 A number of efforts have begun to evaluate urban emissions more rigorously using a variety
71 of methods. For example, detailed inventory-based methodologies can provide high-
72 resolution information on urban emission rates^{12, 15}, with emissions separated by source
73 sectors and spatially and temporally distributed within a city and its surrounds. While these
74 methods have the potential to revolutionize urban emissions information, they require
75 detailed knowledge of all relevant emissions processes and their strengths and have
76 uncertainties that are difficult to quantify.

77 Atmospheric mass balance methods have been used to evaluate urban CO₂ fluxes
78 independent of inventory-based methods. Early work used an aircraft-based mass balance
79 technique to evaluate CO₂ and CH₄ emissions from Indianapolis, and found a standard
80 deviation of 80 % in whole-city CO₂ emission rate estimates for different days measured
81 over a period of roughly one year¹⁶. A slightly different aircraft-based mass balance
82 technique was used to estimate London's CO₂ emission rate, finding a range of about 100 %
83 in their results¹⁷. These studies identified wind speed as the main source of uncertainty
84 along with possible aliasing associated with low sampling frequency. More recent mass
85 balance studies in Indianapolis have achieved improved standard deviation of 20 to 30 %

86 through better meteorological parameter determination¹⁸⁻¹⁹. Uncertainties in the mass
87 balance method have primarily been evaluated by comparing flux estimates for multiple
88 days with one another, using the reasonable assumption that day-to-day variability in the
89 emission rate for individual weekday afternoons over a period of a few weeks is likely to be
90 smaller than the variability induced by the mass balance methodology. Thus the
91 comparison of multiple days provides an assessment of repeatability but not of systematic
92 biases.

93 Another approach is to combine atmospheric ground or tower-based in-situ observations
94 with atmospheric transport modelling. A comparison of simulated CO₂ mole fractions
95 derived from a bottom-up inventory and atmospheric transport model with observed CO₂
96 mole fractions in the Salt Lake City region suggested that this methodology could detect
97 changes in emissions of around 15 %²⁰. Using similar methodologies along with more
98 sophisticated inversion frameworks, several studies have estimated CO₂ emissions for
99 Paris²¹, London²², Rotterdam²³, Boston²⁴, and Indianapolis²⁵. In the Indianapolis inversion
100 study, sensitivities to a range of assumptions embedded within the inversion method were
101 assessed, but the overall uncertainty and potential biases of results were not quantified.

102 Indianapolis is one of the few urban areas where multiple emission assessment methods
103 have been implemented, providing a unique opportunity to compare different methods
104 directly. The Indianapolis Flux Experiment (INFLUX) aims to develop, assess, and minimize
105 uncertainties of methods for quantifying greenhouse gas emissions at the urban scale, using
106 the Indianapolis urban area (Figure 1) as a testbed²⁶. INFLUX goals include determining
107 whole-city emissions of CO₂ and methane, differentiating biogenic and anthropogenic CO₂
108 sources (including source sector allocation), reducing uncertainty in urban emission

109 estimates and, ultimately, providing emission information at 1 km² spatial and weekly
110 temporal resolution. Here we focus on the whole city CO₂ emission rate and compare three
111 different approaches: a science-driven high-resolution urban inventory-based emission data
112 product, an atmospheric transport model inversion based on in situ tower observations, and
113 mass balance flux estimates from aircraft observations. We evaluate differences between
114 the methods, particularly focusing on the use of discrete flask-based measurements to
115 determine fossil fuel CO₂ separately from biogenic CO₂ contributions, and on the
116 contribution of background CO₂ mole fractions to the urban flux estimates. Only winter-time
117 emissions are considered, in order to minimize complications associated with biospheric
118 CO₂ fluxes in this first attempt to compare differences between methods.

119

120 **Whole city emission rate evaluation methods**

121 *Hestia data product*

122 The Hestia high resolution inventory-based data product¹⁵ provides anthropogenic CO₂
123 emission estimates for Indianapolis and the surrounding area (Figure 2). The Hestia
124 approach has now also been implemented in Salt Lake City, Baltimore, and the Los Angeles
125 Basin. It combines multiple data sources that represent a mix of emission-related content
126 and include direct reporting of CO₂ fluxes, reporting of local air pollution (i.e. CO emission
127 reporting), activity data (e.g. traffic counts, aircraft landing/takeoff statistics), fuel
128 consumption statistics, and a variety of sociodemographic statistics. A series of datasets
129 were also used to perform temporal and spatial distribution/downscaling for multiple scales
130 and included data such as building footprints, roads and building occupancy schedules. It
131 includes anthropogenic CO₂ emissions from eight sectors: onroad traffic, offroad vehicles
132 (e.g. construction and farm equipment), railroads, airports, utilities (electricity generation),

133 industry, commercial and residential. Only direct emissions occurring in the domain are
134 considered. There is a small, known contribution of bio-ethanol that is included in gasoline
135 sold in Indiana (10 %, Energy Independence and Security Act of 2007). Mobile emissions
136 comprise 36 % of Hestia’s total emissions, of which 75 % is gasoline combustion¹⁵.
137 Therefore bio-ethanol contributes 3 % of the total Hestia CO₂ emissions and we scale Hestia
138 down by this amount to obtain CO₂ff alone. We assign an initial uncertainty of 12 % to the
139 whole-city Hestia flux, although this is somewhat based on expert judgement since
140 uncertainty is quite difficult to evaluate for this methodology²⁵.

141

142 *Atmospheric inversion*

143 The atmospheric inversion²⁵ utilizes in situ CO₂ observations from 13 towers in and around
144 Indianapolis (Figure 1). The inversion starts with the Hestia “prior” anthropogenic emissions
145 and adjusts the emissions to give the best match with the tower observed CO₂ mole
146 fractions resulting in a posterior flux map for the same domain as Hestia. The posterior flux
147 is resolved spatially (1 x 1 km) and hourly but not by source sector. The analytical solution is
148 calculated for five-day averaged estimates using a Bayesian inversion framework. Only
149 observational data from the afternoon is used (1600 to 2200 UTC, 1100 to 1700 LST), when
150 the model best simulates atmospheric transport. Given the typical transit time of air across
151 Indianapolis of a few hours, the inversion is most sensitive to midday fluxes, is weakly
152 sensitive to early morning fluxes and has very little sensitivity to evening and nighttime
153 fluxes. Since the inversion utilizes CO₂ mole fraction observations, the inversion posterior
154 flux represents the net CO₂ flux including all anthropogenic and biogenic fluxes. To isolate
155 the CO₂ flux inside the domain, the background CO₂ mole fraction is defined by observations
156 from one of the upwind towers (typically either Tower 1 or 9) and adjusted for fluxes that

157 occur upwind of the upwind tower but inside the model domain. Uncertainty was
158 estimated from the one-sigma scatter of 16 different applications of the inverse flux
159 estimate²⁵. This estimate does not include an explicit calculation of uncertainty in
160 atmospheric transport, but earlier work suggests that the atmospheric model used here has
161 modest random error (19° in wind direction and 0.8 m s⁻¹ in wind speed) and small
162 systematic errors (2° in wind direction and 0.1 ms⁻¹) at this time of year²⁷. While the spatial
163 distribution of emissions is affected by these model errors, this has little effect on the
164 aggregated flux over the entire domain²⁷. Since the atmospheric inversion starts with the
165 Hestia emission map as a prior, the inversion results are not entirely independent of the
166 Hestia product. Full details of the atmospheric inversion are described elsewhere²³.
167 In the initial study²⁵, the inversion posterior resulted in 20% higher total emissions than
168 Hestia for the period September 2012 – April 2013 (Table 1). The difference was likely due
169 to that fact that the observational methods calculate total incremental CO₂ including both
170 anthropogenic and biogenic fluxes, whereas the inventory-based product includes only
171 anthropogenic sources²⁸. This explanation is supported by studies demonstrating that
172 although fossil fuel fluxes might be expected to dominate the overall CO₂ emission rate in
173 urban areas, even in the dormant season biogenic CO₂ fluxes increase the total CO₂ flux by
174 10 - 20% in Indianapolis²⁸⁻²⁹ and other cities show a similar pattern³⁰⁻³².

175

176 *Aircraft Mass balance*

177 High resolution in situ CO₂ measurements are made from an aircraft flying downwind of the
178 urban area (varying by day depending on wind direction) at several different altitudes,
179 typically between 12:00 – 16:00 LST (1700 – 2100 UTC). From these, a “curtain” of CO₂ mole
180 fraction observations downwind of the city is developed. Concurrent measurements of

181 wind speed and wind direction allow the CO₂ and CO (also CH₄) emission fluxes from the city
182 to be determined^{16, 18-19}. The emission rate is calculated relative to a background
183 determined from the mole fractions measured in the edges of the downwind transects on
184 the same day. Full details of the method and uncertainties can be found elsewhere¹⁹. Here
185 we only consider the nine flights in November and early December 2014, when we expect
186 that the CO₂ emission rate did not vary substantially from day to day¹⁹ and to avoid the
187 additional complication of strong photosynthetic drawdown in summer. Like the inversion,
188 the aircraft mass balance method evaluates the net CO₂ flux including both anthropogenic
189 and biogenic CO₂ emissions. The aircraft mass balance is independent of both Hestia and
190 the inversion, except for the background correction that is later applied (see later section
191 “Accounting for choice of CO₂ background”). The CO₂ and CO flux estimates use different
192 mole fraction data but are linked through the use of the same wind observations.
193 Investigation of the potential impacts of non-steady atmospheric conditions, and
194 heterogeneous upwind boundary conditions may yield further improvements in the
195 accuracy and precision of aircraft mass balance estimates.

196

197 **Accounting for differences between methods**

198 *Evaluation of CO₂ff vs total CO₂ emissions*

199 Hestia compiles data for anthropogenic CO₂ emissions (from fossil fuel CO₂ and bioethanol
200 combustion). The atmospheric inversion and aircraft mass balance methods both estimate
201 the net total urban enhancement in CO₂, which includes the influence of both
202 anthropogenic and biogenic CO₂ fluxes. To resolve this incompatibility, we use flask
203 measurements of ¹⁴CO₂ to determine the recently added CO₂ff²⁹, along with urban excess

204 (enhancement over background) in CO₂ (CO_{2xs}) and in carbon monoxide (CO_xs) and derive
205 empirical relationships between CO_{2xs}, CO_{2ff} and CO (Table 2).

206

207 Flasks collected in November and December from seven of the 13 towers between 2010 and
208 2016 (151 flasks) and during aircraft flights over and downwind of Indianapolis in November
209 and December of 2010 to 2015 (40 flasks) are included in the analysis. Tower flask samples
210 were collected only during westerly wind conditions so that Tower 1 was upwind of the city
211 (Figure 1) and background was determined from the Tower one measurements. For the
212 aircraft flights, background was defined by a flask measurement collected in the edges of
213 downwind aircraft transects or upwind of the city on the same day. CO_{2ff} was calculated
214 from the observed $\Delta^{14}\text{CO}_2$ and background $\Delta^{14}\text{CO}_2$ in the same flasks²⁹. The CO₂:CO_{2ff} ratio
215 was determined by regressing CO₂ against CO_{2ff} using an ordinary least squares bisector
216 method (Table 2, Fig S1) and the CO:CO_{2ff} ratio was determined in a similar manner (Fig S1).
217 In both cases, the aircraft dataset gives a slightly, but not significantly, higher ratio and
218 reasons for this are discussed in the supplementary material.

219

220 We obtain a CO_{2xs}:CO_{2ff} ratio of 1.1 ± 0.1 ppm/ppm from the flask measurements for the
221 months of November and December, implying that 10 ± 10 % of the CO_{2xs} is due to a local
222 source other than CO_{2ff}. At a continental location such as Indianapolis, this can only be a
223 net biogenic source that is greater than the background biogenic CO₂ fluxes. This is
224 consistent with the known Indianapolis biogenic sources including human and pet
225 respiration, biomass burning (home wood fires and a small power plant in Indianapolis that
226 utilizes biomass¹⁵) and soil respiration. Together, these have been estimated to contribute
227 2,400 to 3,000 mol s⁻¹ of biogenic CO₂ for Indianapolis in winter (13 - 16 % of the Hestia-

228 derived CO₂ff emission rate)²⁸. A similar calculation, but including flask measurements for
229 the months September to April inclusive, gives a slightly higher ratio of 1.2 ± 0.1 ppm/ppm,
230 and is consistent with the September to April biogenic CO₂ flux estimate from a previous
231 study that estimated the wintertime urban biogenic flux, human and pet respiration and
232 known biofuel CO₂ sources within the city²⁸. Thus, we reduce the total CO₂ whole-city
233 emission rate estimates from the atmospheric inversion and the mass balance by a factor of
234 1.1 ± 0.1 to obtain whole-city CO₂ff emission rate estimates from each of these methods
235 (Table S2).

236

237 Samples are also measured for CO, which allows us to obtain a fourth, largely independent,
238 CO₂ff flux estimate that avoids altogether the use of total CO₂ measurements and the
239 associated biogenic CO₂ flux. Instead, in situ CO measurements from the aircraft flights are
240 used in a mass balance calculation analogous to that for CO₂ to determine the whole-city CO
241 emission rate of 108 ± 22 mol s⁻¹¹⁹(Supp Table 2). We then scale whole city CO emission
242 rate by the CO:CO₂ff ratio of 7 ± 2 ppb CO/ppm CO₂ff (Table 2; Fig S1) to determine a CO-
243 based CO₂ff emission rate. Wintertime CO sources in Indianapolis are expected to be almost
244 entirely from fossil fuel combustion, with about 1 % from biomass burning such as home
245 fireplaces^{29, 33-34}, thus no attempt is made to correct for the biomass burning source.

246

247 *Geographic region adjustment*

248 The geographic area for which emissions are evaluated differs among the different
249 methods. Hestia and the atmospheric inversion both evaluate fluxes for the same explicitly
250 defined 87 x 87 km² domain that roughly comprises the nine counties that include and
251 surround the Indianapolis metropolitan area (Figure 1). The aircraft mass balance evaluates

252 fluxes for a (smaller) area that is less well defined and which will differ somewhat for each
253 flight. To compare with the other methods, we define the aircraft footprint, or area of
254 influence, as a box over the city (Figure 2)³⁵. Determined from the location of the downwind
255 flight path and drawing lines upwind the points on either side of the city where the urban
256 emission plume is no longer distinguishable from the regional background CO₂ signal¹⁹. The
257 aircraft footprint varies by flight (Table S2, Figure S2) and is 48 ± 6 % of the full Hestia 87 x
258 87 km² domain, determined from the mean and standard deviation of the fraction of the
259 domain over the nine flights.

260

261 *Time of day and time period of flux assessment*

262 The Hestia CO₂ emission rate estimate reported in Table 1 is the mean flux over all hours of
263 the day. The inversion posterior reports the mean emission rate for the entire diurnal cycle,
264 but incorporates observational data only from the afternoon when the model best
265 simulates atmospheric transport. Hence the inversion is least dependent on the Hestia
266 prior in the midday hours. The aircraft mass balance measurements are always made
267 during the afternoon and are therefore also most sensitive to the daytime fluxes. Thus, we
268 extract only daytime (1600 to 2100 UTC, 11 am to 4 pm local time) emissions from Hestia
269 and the atmospheric inversion so that all methods are comparing approximately the same
270 time of day, and the time of day when the inversion is least dependent on the Hestia prior.
271 The daytime mean emission rate estimates are ~ 20 % higher than the full diurnal estimates
272 for both Hestia and the inversion (Table S2), with the largest difference occurring in the
273 nighttime hours when traffic and commercial sources are quite low¹⁵. There is some
274 variability associated with rush hour and the normal working hours, but choosing a different
275 span of hours between 1200 to 2300 UTC does not significantly impact our results.

276

277 The aircraft flights were all conducted on weekdays, which Hestia predicts have a 13 %
278 higher emission rate than the average across all days of the week, primarily driven by lower
279 emissions on Sundays. In addition, factors such as ambient temperature (and therefore
280 heating demand) and power plant loading result in day to day and seasonal emissions
281 variability. The results previously reported²⁶ for Hestia and the inversion were both for the
282 period of September 2012 to April 2013, whereas the aircraft estimates were from nine
283 flights on weekdays in November and December 2014. Hestia has now been updated for all
284 of 2014, and we therefore subsample Hestia for the same afternoons as the nine aircraft
285 flights. Similarly, the atmospheric inversion has now been extended to November 2014
286 using the same methodology²⁵. We use the atmospheric inversion results for the same
287 seven afternoons as the aircraft flights in November 2014 and estimate the atmospheric
288 inversion results for the two flights on December 1 and 3 2014 as the afternoon flux for the
289 same day of the week, two weeks prior on November 17 and 19, respectively (avoiding the
290 Thanksgiving holiday in the last week of November). Both the inversion posterior and
291 Hestia results show that the mid-afternoon fluxes are quite consistent for each day of the
292 week across the five weeks of November and the first week of December ($\pm 100 \text{ mol s}^{-1}$ one
293 sigma scatter across the five weeks) so this approximation is reasonable.

294

295 *Accounting for choice of CO₂ background*

296 A final factor that must be accounted for is that the Hestia inventory-based methodology
297 sums all known anthropogenic fossil fuel combustion emissions within its domain. In
298 contrast, to isolate the emissions occurring in the Indianapolis domain, the atmospheric
299 methods remove (subtract) the incoming “background” CO₂ signal from the measurements

300 within, or downwind of, the urban area to obtain the CO₂ excess (CO₂xs). In principle, the
301 background CO₂ signal would be the CO₂ mole fraction that would have been measured at
302 the observation site in the absence of the urban emissions. This is an unmeasurable
303 quantity, so in practice, we define the background either as the CO₂ mole fraction at a tower
304 that is immediately upwind of the urban area on a given day (atmospheric inversion) or by
305 linearly interpolating the CO₂ mole fractions measured on the two edges of each downwind
306 transect where the urban emission plume can no longer be discerned (mass balance). We
307 also assume that the background CO₂ mole fraction is consistent spatially (i.e. across all
308 upwind areas) and over the time it takes for air to transit over the city. This is a more
309 reasonable assumption in our wintertime analysis period when the biosphere is dormant.
310 During the growing season the biogenic CO₂ flux is large and varies both due to
311 heterogeneity of land cover, the diurnal cycles of photosynthesis and respiration, and
312 weather patterns.

313

314 In the case of the atmospheric inversion, the CO₂ mole fraction of air entering the model
315 domain is subtracted from the observations before the inversion is performed, so that only
316 the CO₂ emitted within the domain is considered in the inversion posterior. Ideally this
317 background CO₂, or boundary condition, would be known for every point on the model
318 domain boundaries. In practice, it is approximated from observations, larger scale model
319 simulations, or some combination thereof. Urban mesoscale inversions including INFLUX
320 have thus far primarily used upwind tower observations^{20-21, 24-25}. For each day, the
321 background CO₂ signal is determined from the instrumented tower that is upwind of the
322 urban area. A further adjustment is applied to account for the modest CO₂ fluxes that occur
323 within the model domain, but upwind of the tower chosen as background, such that the

324 simulated footprint of the upwind tower on each day is convolved with the prior fluxes from
325 Hestia to obtain a prediction of CO₂xs at the upwind tower relative to the upwind model
326 domain boundary; this is subtracted from the upwind tower CO₂ mole fraction to determine
327 the background expected at the model domain boundary²⁵. The inversion posterior result
328 therefore does not require further adjustment for its fossil fuel background fluxes before
329 comparing to the inventory. This approach assumes that there are no differential influences
330 in upwind versus downwind CO₂ that originate outside of the 87x87 km² inversion domain
331 (e.g. a narrow fossil fuel CO₂ plume that influences the downwind measurements but is not
332 captured at the upwind site).

333

334 In the case of the aircraft mass balance, the measurements from the edges of the downwind
335 transects are used to determine the background CO₂ signal. The edges are defined as the
336 point outside the city where the urban plume can no longer be detected (Figure 2). Yet
337 Hestia predicts that the CO₂ff flux outside the aircraft footprint but inside the Hestia domain
338 is non-zero due to modest emissions from roads and small towns (Figure 1). This can be
339 expected to result in a small but significant increase in the edge CO₂ mole fraction relative
340 to what would have been measured in the absence of those rural emissions (Figure 2). To
341 account for this effect, we determine the mean flux in the rural area outside the aircraft
342 footprint from Hestia, which varies slightly by flight and averages $1.5 \pm 0.2 \text{ mol s}^{-1} \text{ km}^{-2}$ over
343 all flights. We add this Hestia-determined flux per unit area to each gridbox within the
344 aircraft footprint and sum to obtain a background-corrected aircraft mass balance flux
345 (Table S2).

346

347 **Results and discussion**

348 The urban CO₂ emission rates for each method, as first reported in the original papers^{15, 19,}
349 ²⁵, range from 14,600 to 22,400 mol s⁻¹ (Table 1). In aggregate, they have a one-sigma
350 scatter of 21 % and the highest (inversion) and lowest (mass balance) differ by 42 %.
351 Although an improvement over previous uncertainty estimates for urban emissions of 50 –
352 100 % from other studies¹²⁻¹³, the initial INFLUX uncertainties as represented by differences
353 between methods are nevertheless insufficient for detection of emissions trends on the
354 order 10 % per decade.

355

356 Once the differences between methods are accounted for as described in this paper, the
357 spread of values for the different methods of determining the whole city CO₂ff emission rate
358 is 17,700 mol s⁻¹ (CO₂-based mass balance) to 20,500 mol s⁻¹ (Hestia), a difference of 15 %
359 between the minimum and maximum estimates (Table 1, Figure 3). The impact of each
360 adjustment on each method is given in Table S2. In summary, for Hestia the adjusted value
361 is higher than the initial value because subsampling the (higher emission) afternoon period
362 had a larger effect over the smaller footprint area. In contrast, the adjusted value is lower
363 than the initial value for the inversion, where the different time period and the adjustment
364 from total CO₂ to CO₂ff were more important. The adjusted mass balance value was
365 reduced by the conversion from total CO₂ to CO₂ff but increased by the background
366 correction. Overall, each of the adjustments (time period, time of day, day of week,
367 geographic region, CO₂ff vs CO₂, background correction) altered the initial emission rate
368 calculation for that method by -20 to +30 %, with no single adjustment dominating the
369 others.

370

371 The four different realizations of the whole city CO₂ff emission rate from three largely
372 independent methods give a weighted mean emission rate of 19,000 mol s⁻¹. The four
373 methods all agree within their assigned uncertainties and the four mean values have a
374 standard deviation of 1,300 mol s⁻¹ (7 %) at one-sigma (Table 1, Figure 3). We separately
375 calculate a standard error of 1,300 mol s⁻¹ (7 %) from the four realizations, using the
376 uncertainties assigned to each method. Comparison of the standard deviation and standard
377 error can be used to evaluate the appropriateness of the uncertainties assigned to each
378 individual method. When the uncertainties are too large (too small), the standard error will
379 be larger (smaller) than the standard deviation. For our comparison, the consistency
380 between the standard deviation and the standard error imply that the assigned
381 uncertainties of 12 – 18 % (Table 1) for each method are appropriate. We note that the level
382 of uncertainty achieved here requires iteration between top-down and inventory-based
383 methods.

384

385 This study represents the first comprehensive, multiple-method assessment of urban CO₂ff
386 emissions, and the agreement across these demonstrates for the first time CO₂ff emission
387 uncertainty bounds informative for mitigation effort management. We conclude that the
388 methodologies described here can, at least for Indianapolis, be applied collectively to
389 provide emission rates with uncertainties of better than 10 % that will be useful within time
390 frames appropriate to agreed international mitigation approaches/objectives. For example,
391 the Indianapolis City Government aims to be carbon neutral by 2050³⁶. Achieving half of
392 that goal by emission reductions would require a 2 % yr⁻¹ emission rate decrease. Given
393 annual determination of the emission rate with 10 % uncertainty, this trend would be
394 detectable with 95 % confidence in eight years. By increasing the frequency of the emission

395 rate determinations to four times per year, the same trend could be detected with 95 %
396 confidence after only five years, a common time period for reassessment of emissions.

397

398 The analysis presented here is for wintertime, when the biogenic CO₂ flux is small and
399 consistently positive in Indianapolis, with little or no photosynthetic uptake. Relating the
400 total CO₂ measurements used in the atmospheric methods to the fossil fuel CO₂ emission
401 rate will be more challenging during times when the biosphere is more active with large and
402 varying biogenic CO₂ fluxes both within and around the urban area. One path forward is to
403 expand the use of the combined flask ¹⁴CO₂ and in situ CO measurements to evaluate the
404 CO₂ff emission rate^{30, 37-38}. Other ancillary anthropogenic trace gases may be worth
405 investigating³⁹⁻⁴⁰. Improved control on the biogenic CO₂ fluxes inside and outside the city is
406 an area of active research through both modelling^{24, 31} and measurement that will likely
407 yield significant improvements.

408

409 The greatly improved agreement between methods suggests that any of these methods
410 could be employed alone to evaluate urban emissions, although the uncertainties and
411 biases in each method could vary depending on the characteristics of any individual city⁴¹.
412 The inventory-based data product offers detailed emission maps and process information,
413 yet the large data-gathering effort required means it will be more practical in some cities
414 than others. An atmospheric inversion based on long-term observations provides the
415 opportunity to evaluate changes in emissions through time and has been shown to give
416 robust results even with simpler prior flux estimates such as the ODIAC product^{25, 42}, but
417 requires a long-term commitment to measurement infrastructure. The aircraft mass balance
418 method could more quickly provide emission rate estimates for a suite of cities using a

419 single instrumented aircraft, albeit with limited time resolution for each city. Importantly,
420 use of any single method will continue to limit the ability to assess methodological bias and
421 uncertainty.

422

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583 **Tables**

584

585 Table 1. Methodologies and results for each CO₂ emission rate determination and the
 586 weighted mean and standard error of the four different methods. Initial CO₂ emission rates
 587 are the values and one sigma uncertainties reported in previous publications^{15, 19, 25-26}.
 588 Adjusted CO₂ff emission rates are the values determined with the adjustments described in
 589 the text and representing the CO₂ff emission rate for nine days in November to December
 590 2014, 11 to 16 LST, for the region representing the aircraft footprints, and the mass balance
 591 corrected for background. Uncertainties are based on the one sigma scatter of results for
 592 each of the nine flights. See supplementary material for individual flight values.

Method	Includes...	Domain	Time of day	Time period	Initial CO ₂ (mol/s)	Adjusted CO ₂ ff (mol/s)
Hestia inventory-based	Fossil CO ₂ + bioethanol	9 counties	All	Sep 2012 – Apr 2013	18,300 ± 2,200	20,500 ± 2,400
Inversion/tower CO ₂	Total CO ₂	9 counties	All	Sep 2012 – Apr 2013	22,400 ± 500	18,200 ± 2,100
CO ₂ -based mass balance	Total CO ₂	Aircraft footprint	Mid-afternoon	Nov – Dec 2014	14,600 ± 3,300	17,700 ± 3,200
CO-based mass balance	Fossil CO ₂	Aircraft footprint	Mid-afternoon	Nov – Dec 2014	NA	19,800 ± 3,400
Weighted mean	Fossil CO₂	Aircraft footprint	Mid-afternoon	Nov – Dec 2014		19,000 ± 1,300

593

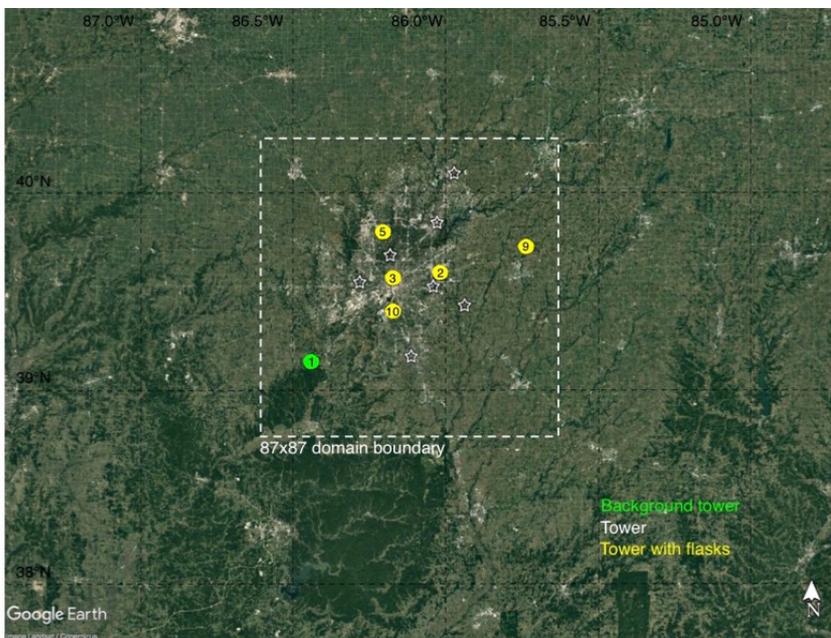
594 Table 2. Emission ratios for CO₂xs:CO₂ff and COxs:CO₂ff determined from flask samples
 595 collected from towers and aircraft from 2010 to 2016. Ratios and their one sigma
 596 uncertainty are determined from the correlation between CO₂ff and CO₂xs or COxs, using
 597 ordinary least squares bisector regression (Table S1; Figure S1).

	CO₂xs:CO₂ff (ppm/ppm) (n, r²)	COxs:CO₂ff (ppb/ppm) (n, r²)
All Nov – Dec	1.1 ± 0.1 (186, 0.8)	7 ± 2 (191, 0.6)
Towers Nov - Dec	1.1 ± 0.1 (151, 0.8)	7 ± 2 (151, 0.6)
Aircraft Nov - Dec	1.2 ± 0.1 (35, 0.9)	9 ± 2 (40, 0.7)
All Jan - Oct	NA	8 ± 1 (788, 0.5)

598

599 **Figures**

600

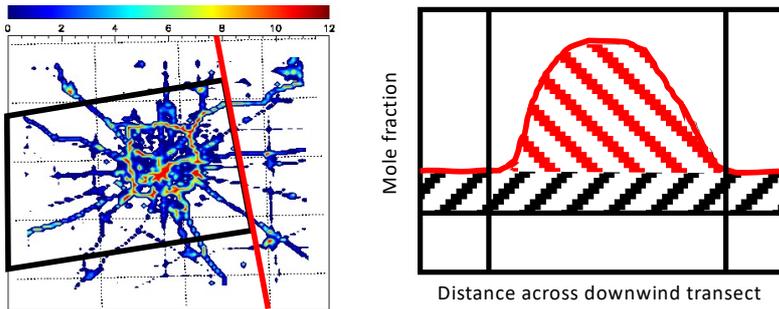


601

602 Figure 1. Map of Indianapolis region showing Hestia/inversion domain and the sampling

603 towers.

604



605

606 Figure 2. Left: Example of how the footprint area is determined for an aircraft mass-balance

607 transect measurement, defined by the box enclosed by the black and red lines. For context,

608 the footprint area is overlaid on the Hestia spatial distribution of emissions. Right:

609 Schematic of CO₂ measurements for a single downwind transect. The solid red line indicates

610 the downwind flight path where the CO₂ and CO measurements are made and used to

611 determine the mass balance emission rate from the urban area. The edges of the urban

612 plume are defined as the point where the urban CO₂ plume is not distinguishable, indicated

613 by the vertical black lines in the right panel. To determine the footprint, lines are projected

614 upwind from the edge points in the direction of the wind to the upwind edge of the Hestia

615 domain as shown by the black box in the left panel. The effect of rural CO₂ff emissions is

616 also shown in the right panel. If the CO₂ff flux in the rural area is zero, then the urban CO₂ff

617 flux is defined by the red hatched area under the red curve. However, in the case that rural

618 CO₂ff emissions are small but non-zero, the CO₂ mole fraction may not appear to vary in the

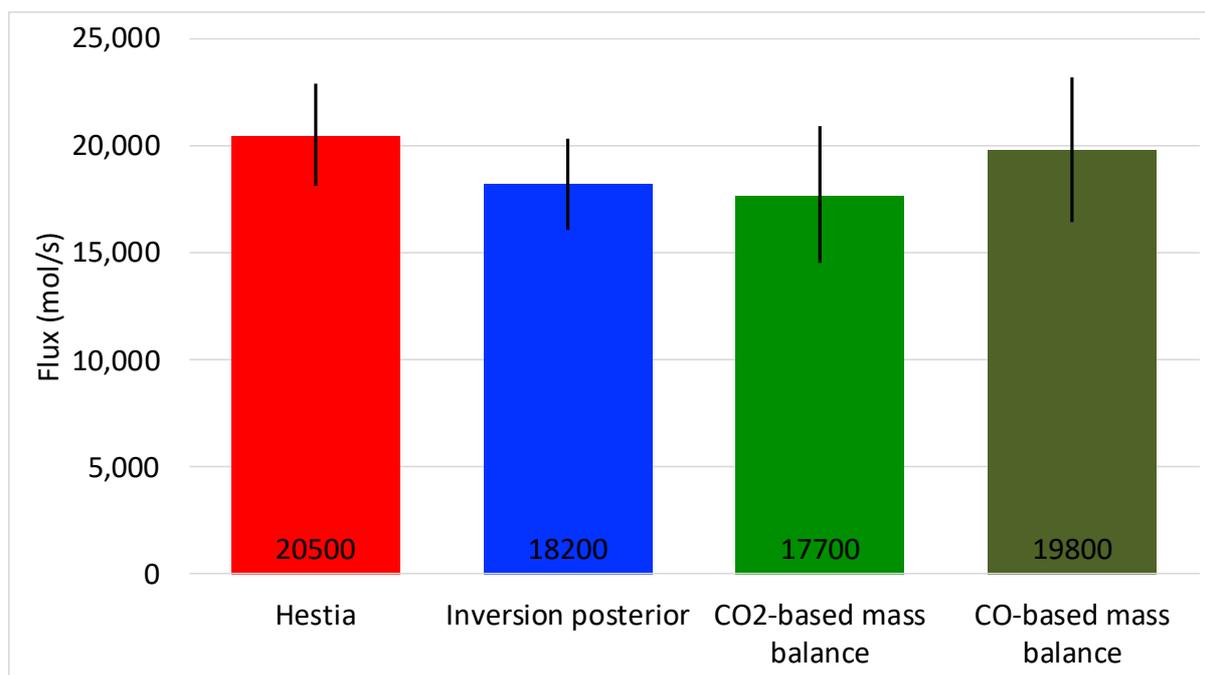
619 edges but will still be elevated relative to what it would have been in the absence of those

620 rural emissions, shown as the black hatched area. Thus, when rural emissions occur upwind

621 of the edges of the downwind aircraft transect, the aircraft mass balance method will

622 underestimate total urban emissions.

623



624

625 Figure 3. Adjusted CO₂ff emission rates. Values were determined with the adjustments
626 described in the text and representing the CO₂ff emission rate for nine days in November to
627 December 2014, midday hours only, for the region representing the aircraft footprints, and
628 the mass balance corrected for background. Error bars are the one sigma scatter of the
629 results for each of nine flights.

1 **Supplementary Material**

2

3 **Supplementary Table 1.**

4 Emission ratios for individual towers, aircraft and different time periods. The ratios
5 determined for only November and December 2014 are not significantly different than
6 those determined for all years 2010 – 2016, although the smaller dataset for the single year
7 results in a larger uncertainty for CO_xs:CO₂ff. Defining “winter” as the months of September
8 – April inclusive, as was used in previous INFLUX publications¹⁻² results in a slightly higher
9 CO₂x_s:CO₂ff ratio and a somewhat weaker correlation. This is likely due to stronger biogenic
10 CO₂ fluxes in the fall and spring months than in the colder, lower daylight hour months.
11 There is no difference in the CO_xs:CO₂ff ratio for the longer winter period and larger
12 dataset.

13

14 CO_xs:CO₂ff varies substantially between towers, and the aircraft result is slightly higher than
15 the tower average. It is likely that this is related to sampling biases. The 151 November –
16 December tower samples are from five different towers (supplementary material), and
17 CO:CO₂ff ratio varies by tower, with the highest ratio (10 ± 2 ppb/ppm) observed at Tower
18 two and the lowest at Tower ten (2 ± 1 ppb/ppm). This is consistent with Tower two being
19 most influenced by traffic (with high CO:CO₂ff emission ratio) and Tower ten being strongly
20 influenced by power plant emissions with low CO:CO₂ff emission ratio. Although five tower
21 locations are not sufficient to perfectly observe the entire urban emissions, the mix of sites
22 might be expected to give a reasonable approximation of overall emissions. In each aircraft
23 flight, the limited number of flasks are deliberately collected in the urban plume, but
24 outside the obvious (higher CO₂) power plant plume. Since the power plant has a very low

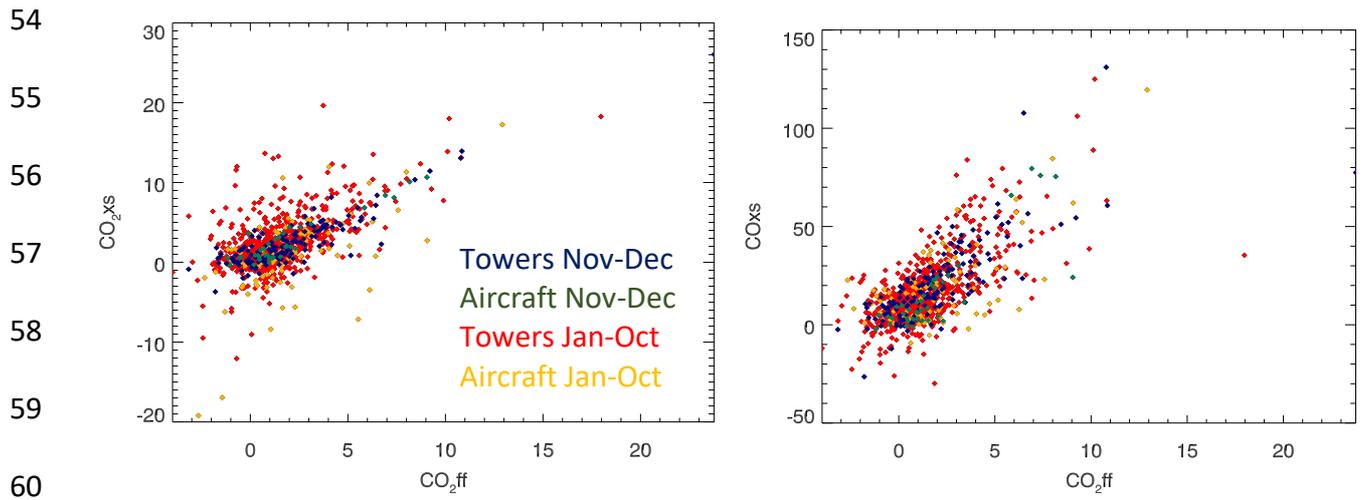
25 CO:CO_{2ff} emission ratio³, this can be expected to bias the aircraft samples to a higher
 26 CO:CO_{2ff} ratio. Note that the aircraft in situ CO₂ and CO measurements are taken during the
 27 entire flight, including the power plant plume, so this applies only to the flask
 28 determinations. For CO:CO_{2ff}, there is no significant difference between winter and
 29 summer ratios (8 ± 1 for all 788 flasks), although other evidence⁴ suggests that CO:CO_{2ff}
 30 should be slightly higher in summer due to production of CO from volatile organic
 31 compounds.

	Slope CO₂xs:CO_{2ff} (ppm/ppm) (n, r²)	Slope COxs:CO_{2ff} (ppb/ppm) (n, r²)
All Nov – Dec	1.1 ± 0.1 (186, 0.8)	7 ± 2 (191, 0.6)
Towers Nov - Dec	1.1 ± 0.1 (151, 0.8)	7 ± 2 (151, 0.6)
Tower 2 Nov - Dec	1.1 ± 0.1 (37, 0.7)	10 ± 2 (37, 0.8)
Tower 3 Nov - Dec	1.1 ± 0.1 (34, 0.9)	5 ± 2 (34, 0.5)
Tower 5 Nov - Dec	1.1 ± 0.1 (29, 0.6)	8 ± 2 (29, 0.7)
Tower 9 Nov - Dec	1.2 ± 0.1 (33, 0.6)	6 ± 1 (33, 0.5)
Tower 10 Nov - Dec	1.1 ± 0.1 (9, 0.8)	2 ± 1 (9, 0.7)
Aircraft Nov - Dec	1.2 ± 0.1 (35, 0.9)	9 ± 2 (40, 0.7)
All Nov – Dec 2014 only	1.1 ± 0.1 (46, 0.8)	6 ± 3 (51, 0.5)
All Nov - Apr	1.2 ± 0.1 (472, 0.7)	7 ± 1 (476, 0.5)
All Sep - Apr	1.2 ± 0.1 (648, 0.6)	7 ± 1 (652, 0.5)
All Jan - Oct	NA	8 ± 1 (788, 0.5)
Towers Jan - Oct	NA	8 ± 1 (699, 0.5)
Aircraft Jan - Oct	NA	8 ± 1 (89, 0.5)

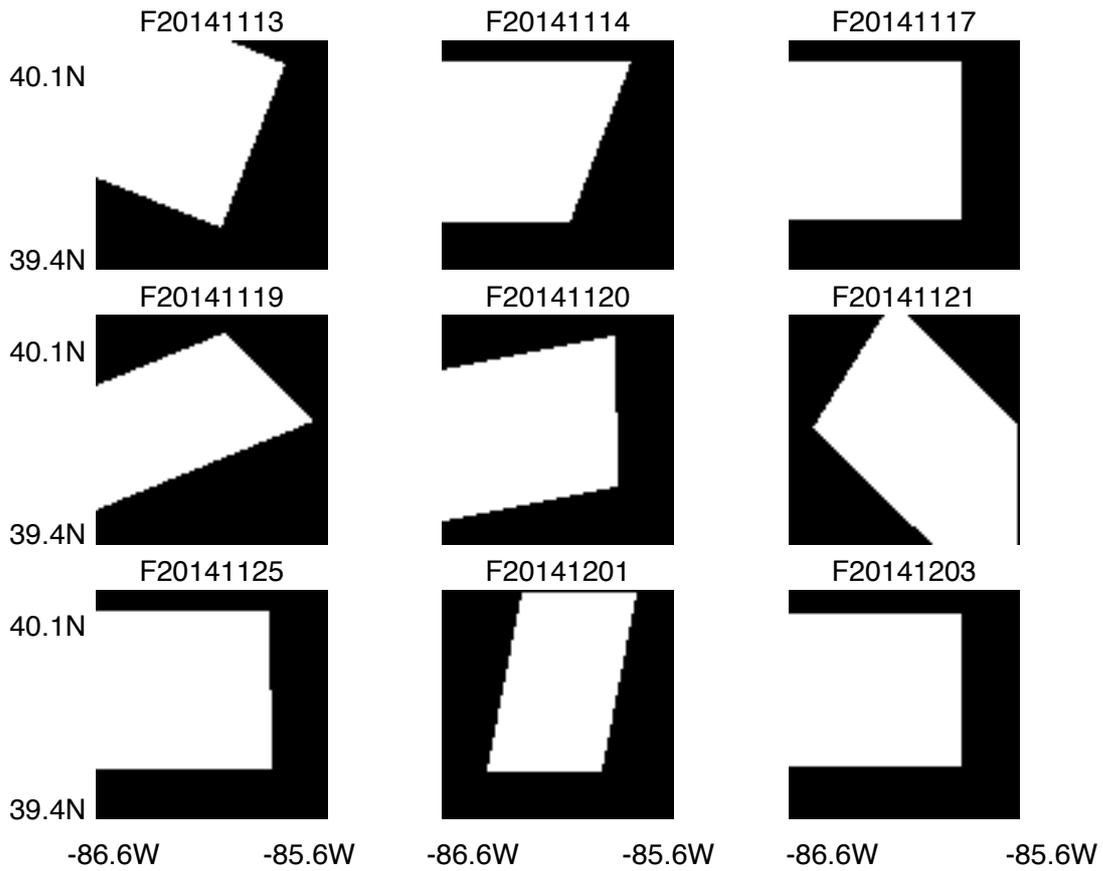
32

33 **Supplementary Table 2.**

34 Emission rates and other calculated values for each of the nine aircraft flights. The fraction
35 of 87x87 km Hestia domain that is inside the aircraft footprint is given as % area in footprint
36 (maps of the footprints are given in Supplementary Figure 2. The Initial CO₂ and CO₂
37 emission rates are given for the aircraft, along with CO₂ff calculated using emission ratios
38 CO₂xs:CO₂ff of 1.1 ± 0.1 ppm/ppm and CO_xs:CO₂ff of 7 ± 2 ppb/ppm. Background corrected
39 aircraft emission rates add the background correction for mass balance determined from
40 Hestia, which is calculated by adding the CO₂ flux rate in the edges (outside the aircraft
41 footprint) to every 1x1 km gridbox inside the aircraft footprint and summing across all
42 gridboxes (see main text for explanation). Hestia 2014 uses the 1 km resolution 87x87 km
43 gridded Hestia product for 2014. The CO₂ emission rate for the footprint is determined by
44 subsampling for the aircraft footprint for that flight day, and averaging across the hours
45 1600 to 2100 UTC. CO₂ff emission rate in footprint is determined by scaling the CO₂
46 emission rate down by 3.5% to remove the bioethanol component. The CO₂ flux rate in the
47 footprint and in edges are the mean flux rate in mol/s/km² for all gridboxes inside or outside
48 the footprint, respectively. The fraction of the total emissions in the domain during that
49 time that occur in the aircraft footprint is given as % emissions in footprint. The
50 atmospheric inversion is reported at 1 km resolution on the same 87x87 km grid, and is
51 sampled in the same manner as Hestia. The CO₂ff emission rates used in the comparison in
52 the main text are shown in bold.



61 **Supplementary Figure 1.** Flask measurements of CO₂ff vs CO₂xs (left) and COxs (right) used
62 to determine the ratios CO₂xs:CO₂ff and COxs:CO₂ff. Units are ppm for CO₂ff and CO₂xs and
63 ppb for COxs. Ratios were determined using an ordinary least squares bisector method^{3,5}.



64

65 **Supplementary Figure 2.** Footprints for each of the nine flights. Flight names are listed as

66 Fyyyymmdd. White area is the aircraft footprint and the black area is the area inside the

67 domain but outside the aircraft footprint.

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71 **Supplementary References**

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